

1 **Rock abundance on the lunar mare on surfaces of different age: Implications for**
2 **regolith evolution and thickness**

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17 **Key Points:**

- 18 • Crater frequency and age of mare units are inversely correlated with rock abundance,
19 consistent with regolith thickening over time.
- 20 • On old surfaces with thick regolith, the median rock abundance is non-zero, consistent
21 with re-excavation of rocks from within the regolith.
- 22 • The half-life of meter-scale surface rocks as derived in the work presented here is 80 ± 20
23 Myr.

24

25 Abstract

26 The growth of lunar regolith over time affects surface rock abundance, because larger, less
27 frequent impacts are needed to penetrate thicker regolith developed on older surfaces and
28 excavate rocks. On younger surfaces with thinner regolith, smaller, more frequent impacts are
29 sufficient to excavate rocks. We quantify the correlation between observed rock abundances and
30 age on the lunar surface by comparing Diviner rock abundance data to the surface ages of inter-
31 crater parts on the maria. Our observations show the expected negative correlation between age
32 and rock abundance. The commonality of non-zero rock abundance values on ancient surfaces,
33 combined with a simple Monte Carlo model of the rock excavation process, suggest that rocks
34 re-excavated from the regolith volume contribute to the presently observed rock population on
35 the lunar surface. The half-life of meter-scale surface rocks most consistent with our
36 observations is 80 ± 20 Myr.

37

38 Plain Language Summary

39 The lunar surface is covered by regolith, which includes particles ranging in size from dust to
40 boulders. Past work has suggested that regolith is meters to tens-of-meters thick, with average
41 thickness increasing with terrain age. At least in the maria, beneath the regolith is fragmental
42 bedrock. When impacts large enough to penetrate the regolith occur, boulders from this
43 underlying bedrock are excavated onto the surface. Excavating boulders is expected to be more
44 difficult where the regolith is thicker, because thicker regolith limits the ability of craters to eject
45 material from the underlying fragmental bedrock. This implies larger and rarer impacts are
46 needed to excavate rocks on older terrain. We tested this idea by comparing the rock abundance
47 data from the Diviner Lunar Radiometer Experiment on the Lunar Reconnaissance Orbiter
48 spacecraft with previously obtained crater statistics for part of the maria, excluding the area
49 within and immediately adjacent to $D \geq 1$ km craters. We show that older maria have, on average,
50 lower fractional rock abundances, consistent with a thicker regolith on older surfaces influencing
51 rock excavation. Rocks excavated from within the regolith are necessary to be consistent with
52 the continued presence of non-zero rock abundance on old terrains.

53 1. Introduction

54 The surface of the Moon is covered by regolith, a layer of poorly sorted particles ranging
55 in size from dust to boulders, the median particle size of which is very fine sand (Carrier et al.,
56 1991). Because of its ubiquity, the regolith is primarily what we observe on the Moon with
57 remote sensing instruments and is the material we interact with when exploring the lunar surface.
58 Many of the important geological, geochemical, mineralogical, and geotechnical characteristics
59 that define lunar regolith were established by the Apollo missions and its precursors (e.g.,
60 McKay et al., 1991). Modern remote sensing methods allow us to extrapolate regolith
61 characteristics from landing sites to places that have not yet been explored in situ.

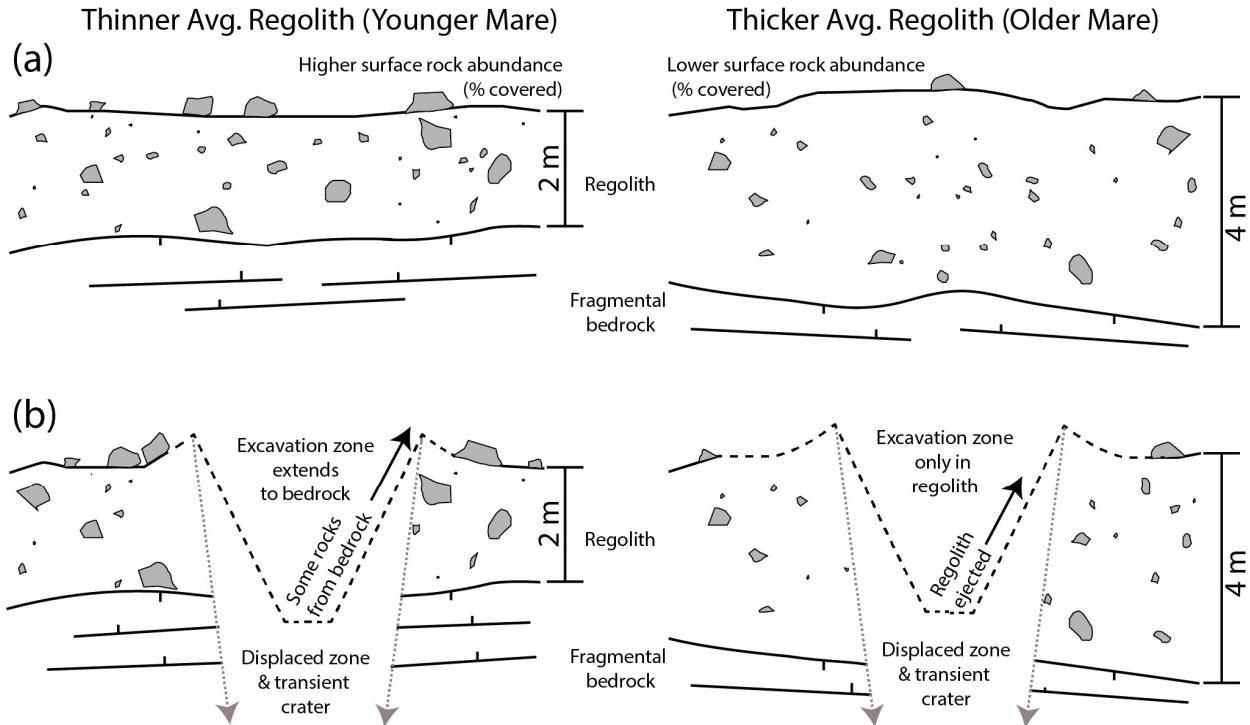
62 The existing paradigm for regolith growth and evolution is that it is dominated by impact
63 cratering and gardening (e.g., Shoemaker et al., 1967; McKay et al., 1991; Costello et al., 2018).
64 Because of the similarity between impacts and explosions, this can be thought of as an explosive
65 demolition process. Using the Neukum production function (NPF) for the Moon combined with
66 scaling calculations to determine the size of impactors (Ivanov, 2001), the kinetic energy
67 delivered by impacts over the last 3 Ga that formed $10 \text{ m} \leq D \leq 500 \text{ m}$ craters is $\sim 3\text{--}6 \times 10^{15}$
68 J/km^2 , approximately equivalent to a megaton of TNT/ km^2 , with $\sim 10^4$ unique events/ km^2 . This
69 demolition process is why there is almost no bedrock exposed on the Moon's upper surface, the
70 median grain size of the surface has been reduced to very fine sand, and regolith thickens with
71 time.

72 Several of the instruments on the Lunar Reconnaissance Orbiter (LRO) have provided
73 new data relevant to rock populations and regolith evolution, particularly the Lunar
74 Reconnaissance Orbiter Camera (LROC) (Robinson et al., 2010), the Diviner Lunar Radiometer
75 Experiment (Paige et al., 2010; Hayne et al., 2017), and Miniature Radio Frequency (Mini-RF)
76 radar (e.g., Nozette et al., 2010; Raney et al., 2011; Cahill et al., 2014). Many analyses of these
77 data have focused on the rock abundance in craters' ejecta (Ghent et al., 2014; Fassett et al.,
78 2018; Mazrouei et al., 2019; Nypaver et al., 2021), rather than in the inter-crater regions of the
79 maria. Additionally, the abundance of meter-scale rocks has also been manually assessed with
80 LROC in a few selected areas of particular interest, such as landing sites (Li et al., 2017; Wu et
81 al., 2018; Watkins et al., 2019), as well as on a more global basis with machine learning (Bickel
82 et al., 2020). By looking at craters of different ages, LROC imaging allowed an estimate of the

83 half-life of rocks (≥ 2 m) as 40–80 Myr (e.g., Basilevsky et al., 2013). The half-life here is
84 defined as the period where 50% of simultaneously exposed rocks in a population would be
85 destroyed; or, equivalently, the period that any given rock has a 50% chance of survival after
86 exposure. This short lifetime implies that rocks on the lunar surface are very susceptible to
87 destruction by impacts (Hörz et al., 2020) and by other processes such as thermal fatigue (e.g.,
88 Molaro et al., 2017). However, craters and landforms older than 300 Myr can maintain an excess
89 of nearby meter-scale rocks for much longer than the lifetime of individual rocks (Ghent et al.,
90 2014; Bickel et al., 2020; Nypaver et al., 2021). Two factors may contribute to this persistent
91 excess of rocks. First, regolith mobility in areas of substantial topographic relief can allow
92 exhumation of rocks from the subsurface. Second, around the D=18-90 km craters considered by
93 Ghent et al. (2014), the initial excavated rock population likely includes rocks much bigger (10m
94 and above) than those looked at by Basilevsky et al. (2013) that may take additional time to be
95 destroyed.

96 Past work has suggested that bedrock is converted to regolith on the Moon at a rate of
97 order $\sim 1\text{--}2\text{m/Gyr}$ (Langevin and Arnold, 1975; Hörz et al., 1991; Xie et al., 2021) so that
98 regolith thickness correlates reasonably well with unit surface age (e.g., Shkuratov and
99 Bondarenko, 2001; Fa et al., 2014). However, there are three subtleties regarding the thickening
100 of regolith with time, evident from Monte Carlo modeling (e.g., Oberbeck et al., 1973; Quaide
101 and Oberbeck, 1975) and theory (Hirabayashi et al., 2018). First, the actual rate of regolith
102 growth is faster when and where the regolith is thin; as a result, regolith growth slows with time
103 (Oberbeck and Quaide, 1968). This is because smaller craters that form more frequently can only
104 create new regolith when the depth to underlying fragmental bedrock is sufficiently thin (e.g.,
105 Fig. 1).

106



107
108 Figure 1. The hypothesis for this paper is that (a) younger areas will have thinner average regolith and more rocks at
109 the surface. As illustrated in (b), a small crater (~10–20 m in diameter, vertical exaggeration ~10×) can excavate
110 through a thinner regolith layer but not a thicker regolith area. On average, this makes ejected blocks from
111 underlying bedrock more common on younger, thinner regolith areas.

112
113 Second, even in a local area of a given age, i.e., on an individual mare flow unit, there is
114 expected to be an appreciable range in regolith thickness (e.g., Oberbeck and Quaide, 1968;
115 Wilcox et al., 2005; Bart et al., 2011; Hirabayashi et al., 2018; Yue et al., 2019), so describing
116 only the central tendency (median or average) of regolith thickness neglects a substantial amount
117 of variability. The reason for this intra-unit variability is that regolith thickness is highly
118 sensitive to the size and distance of the nearby largest craters ejecting regolith or brecciating
119 material on their interior (e.g., Hirabayashi et al., 2018). As a result of this variability, the
120 distribution of regolith thickness around the median or expected value for a given surface is not
121 symmetric. A small fraction of area on a local terrain can have much thicker regolith than normal
122 (perhaps several multiples of the median), whereas the thinnest part of the regolith is closer to
123 the median (see, e.g., Hirabayashi et al. 2018; their Fig. 10).

124 Third, we might expect the rate of regolith thickening was much higher early in the
125 history of the Moon, because the impact rate was higher. For example, with the NPF flux, an
126 equivalent number of craters are expected to be produced in the last 1 Ga as in 100 Ma between

127 3.37 and 3.47 Ga. Since the bulk of the maria were emplaced >3.3 Ga (e.g., Hiesinger et al.,
128 2011), the regolith on the maria may predominantly be expected to have mostly formed in early
129 time periods and been augmented by slower growth rates in the last billion years. While this
130 general trend may remain correct, regolith thickness evolution may be more complicated if there
131 was a shift in the shape of the impactor population's size-frequency distribution along with the
132 increased flux, as recently suggested by Xie et al. (2021).

133 The hypothesis of this study is that the abundance of rocks in low-relief areas of the
134 maria should be sensitive to regolith thickness, and thus surface age (see Fig. 1; also note that an
135 inverse correlation between the abundance of rocks and regolith thickness has been suggested at
136 the Surveyor landing sites (e.g., Shoemaker and Morris, 1970)). The basic reason for this
137 expected sensitivity is that the rock population of the lunar surface is controlled by how many
138 rock-excavating craters can form over the comparatively short lifetime of surface rocks (i.e., Fig.
139 1). Where regolith is thinner, smaller, more frequent impacts can excavate rocks; where regolith
140 is thicker, larger, less frequently occurring impacts are needed. The goal of this study is to
141 evaluate the correlation between rock abundance and surface age and use these observations to
142 refine the model of regolith evolution.

143 We address this hypothesis using observations from Diviner rock abundance data
144 (Bandfield et al., 2011) combined with mapped crater density information from Fassett and
145 Thomson (2014). In the discussion, we also describe a simple Monte Carlo model of lunar
146 surface rock abundance that we have used to better understand the observations.

147 **2. Data and Methods**

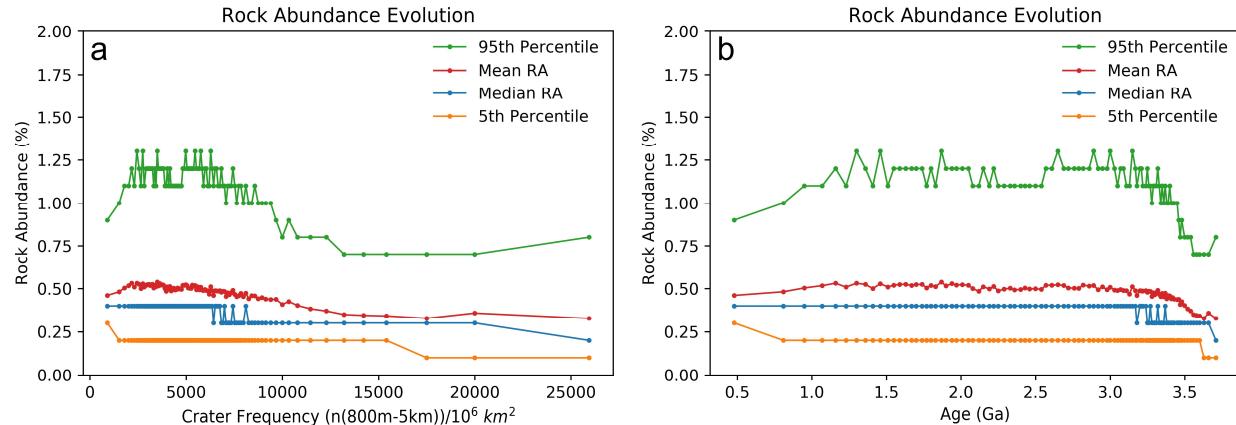
148 For this project, the main dataset is lunar rock abundance derived from Diviner
149 observations (Bandfield et al., 2011; 2017). This Diviner rock abundance map was derived by
150 fitting the measured nighttime thermal radiance with a two-component model assuming the
151 observed scene is a mixture of fine-grained material and rock. The resulting map pixels represent
152 the areal fraction of the surface covered by rock fragments roughly ≥ 1 m (cf. Bandfield et al.,
153 2017). Diviner-derived rock abundances have been validated with what is observed in LROC
154 data (Bandfield et al., 2011), and LROC-derived rock abundance appears to agree reasonably
155 with what is estimated at landing sites (e.g., Watkins et al., 2019).

156 To estimate unit surface ages, we rely on measurements of surface crater densities from
157 Fassett and Thomson (2014). They mapped 13,657 craters in the D=800 m to 5 km size range on
158 a portion of the mare (see their Fig. 1), selected to exclude areas near very large, post-mare
159 craters ($D \geq 20$ km) as well as to manually exclude areas with chains of obvious secondaries.
160 Fassett and Thomson (2014) generated a crater density map of these craters in 50 km-radius
161 moving neighborhoods at 5 km spatial resolution (hereafter referred to as neighborhood crater
162 frequency) (see their Fig. 5a). An advantage of this type of moving neighborhood method (e.g.,
163 Ostrach and Robinson, 2014) is that it is strictly based on mapped superposed craters, so it is not
164 reliant on other geologic inputs to derive crater statistical information. For analysis, these crater
165 frequencies are converted to ages using the NPF chronology model (Ivanov, 2001) from the
166 formation rate of D=800 m-5 km craters (see supplement text S1). The systematic uncertainty in
167 our ages thus inherits uncertainties underlying the NPF chronology (e.g., complexities related to
168 landing site geology, secondary contamination, etc).

169 Diviner rock abundance pixels were extracted and assigned their corresponding
170 neighborhood crater frequency. To focus our dataset primarily on inter-crater regions, we also
171 excluded the rock abundance/crater frequency pairs for the portions of the maria within one
172 crater diameter of the rim of $D \geq 1$ km craters based on the crater catalog of Robbins (2019). This
173 does not exclude the contribution of these $D \geq 1$ km craters to the neighborhood crater frequencies
174 or ages, but does exclude them from the analysis of how the rock abundance evolves. The
175 resulting total study area was 1.54×10^6 km², approximately 25% of the maria, incorporating
176 33.2×10^6 rock abundance pixels. The data from these regions were sorted by increasing
177 neighborhood crater frequency and grouped into 100 bins using python library pandas' qcut
178 function. This function bins the Diviner data into constant-size groups (3.3×10^5 per bin) based on
179 frequency. The reason this statistical approach is necessary is that rock abundances are expected
180 to be highly stochastic: even for units of the same age and regolith thickness, the rock abundance
181 should be expected to vary significantly based on the size and recency of rock-excavating
182 craters. Binning together a large number of pixels with consistent crater frequency (or,
183 equivalently, a large amount of area of equivalent age), this effect can be minimized, and overall
184 trends revealed. The distribution of rock abundance values at a given locale was then assessed by
185 extracting the 5th-, 50th- (median), mean, and 95th-percentile of the rock abundance values in
186 each age bin-defined measurement area. The median frequencies for these groups are equivalent

187 in the NPF chronology to expected crater population accumulated on surfaces over ~0.5 Ga to
 188 3.7 Ga.

189



190

191 Figure 2. (a) Results for Diviner rock abundance (y-axis) vs. neighborhood crater frequency (x-axis) on
 192 surfaces in 50-km radius moving neighborhoods. (b) Results for Diviner rock abundance (y-axis) vs
 193 Neukum model age in billions of years (x-axis). The mean rock abundance slowly declines with
 194 increasing crater frequency / age. Because craters accumulated faster before 3 Ga, the decline in rock
 195 abundance is most prominent on the oldest surfaces (see Figure S1 in the supporting information for a
 196 detail of (b) from 3.0 to 3.75 Ga).

197

198 3. Results

199 The observed median, mean, 5th-percentile, and 95th-percentile rock abundances on the
 200 maria are shown as a function of neighborhood crater frequency in Figure 2a, and as equivalent
 201 model ages in Figure 2b. The model ages in Figure 2b come from the same data as 2a, with
 202 the x-axis rescaled to convert the neighborhood crater frequencies to ages using the NPF
 203 chronology.

204 Overall, the median and 5th-percentile of rock abundances show a nearly monotonic
 205 decline in rock abundance as a function of neighborhood crater frequency in Fig. 2a. The
 206 observed decline is nonlinear because it convolves several underlying complex processes (i.e.,
 207 the cratering and regolith growth rates). The oscillatory behavior of the median for $n(0.8\text{--}5 \text{ km})$
 208 $\sim 6000\text{--}8000 \text{ craters}/10^6 \text{ km}^2$ (ages $\sim 3.1\text{--}3.4 \text{ Ga}$) is due to the fact that the Planetary Data System-
 209 released Diviner fractional rock abundances values were stored with precision of 0.001 (0.1%),
 210 so all the percentile values match this fixed level of precision as well. The mean rock abundances

211 given in Fig. 2 are less affected by this fixed precision but are more sensitive to the rockiest
212 fraction on surfaces of a given age than the median.

213 As a function of age (Fig. 2b), the median rock abundance is observed to be essentially
214 constant for surfaces <3.1 Ga at 0.4%, decreases for surfaces >3.1 Ga to 0.3%, and to 0.2% at 3.7
215 Ga. The 5th-percentile (least rocky part of surfaces of a given age) declines from 0.3% in the
216 youngest bin (~ 0.44 Ga) to 0.2% from ~ 0.77 – 3.6 Ga, to 0.1% at 3.7 Ga. Many of the least rocky
217 areas of the Moon still have non-zero rock abundance (Bandfield et al., 2011), even in the
218 highlands, and our 5th-percentile values here are not exceptional in never reaching 0.0%. The
219 95th-percentile (rockiest part of surfaces of a given age) initially increases in rock abundance,
220 from 0.9% in the youngest bin, to 1.2% from ~ 1.2 Ga to ~ 3.3 Ga, then declines back to 0.7–0.8%
221 $>\sim 3.5$ Ga. The decreasing rock abundance trend in Fig. 2 for all series are statistically significant
222 using the Mann-Kendall test (Hussain and Mahmud, 2019), as is the decline in rock abundance
223 with frequency and age in the raw dataset as a whole (see supporting Text S2). However, this
224 relationship is noisy and only holds when analyzed statistically or aggregated over large surface
225 areas. For a given location, the rock abundance is not a useful predictor of age since it is
226 contingent on the specific cratering history at that location.

227 The first five bins with the lowest nearby crater density (i.e., youngest surroundings, <1
228 Ga-equivalent model age) in Fig. 2 are anomalous in two respects. First, the 95th-percentile and
229 mean rock abundance increase with age (or neighborhood crater frequency) in those bins, unlike
230 the behavior of the rest of the data. Second, the equivalent model ages for these surfaces are
231 lower than expected, given either (a) the age of mare basalts in existing sample collections or (b)
232 the inferred age of units when measured at larger scales than 50-km. The most likely source for
233 this anomaly is small number statistics, i.e., a 50 km radius was not enough to adequately sample
234 the density of craters on sparsely cratered and relatively young terrains. In other words, these
235 observed low neighborhood frequencies may be the stochastically less cratered fraction of older
236 geologic units, rather than broad surfaces that are truly geologically young. It is thus somewhat
237 uncertain how robust the behavior we see in these lowest neighborhood frequency/youngest age
238 bins is.

239 Excluding the five bins with the lowest neighborhood crater frequency, the 95th-
240 percentile rock abundance (the rockiest part of the surface for a given age/frequency) declines

241 more quickly than the median rock abundance, which declines more than the 5th-percentile. In
242 other words, we see the rockiest fraction of old terrain getting less rocky faster than its typical
243 rock abundance.

244 **4. Discussion, Comparison to Earlier Work, and Summary**

245 **4.1 Observed evolution of rock abundance with time**

246 The fact that rock abundance statistics are all strongly correlated with neighborhood
247 crater frequency (and thus age), and decline nearly smoothly for surfaces >1 Ga, supports the
248 interpretation that regolith thickness is an important factor in controlling the observed rockiness
249 of its surface. Additionally, the trends in rock abundance we observe with time, especially the
250 much faster decline in rock abundance on surfaces older than ~3.1 Ga, are consistent with the
251 exponentially higher impact flux early in lunar history that helped especially thicken the regolith
252 on older terrains.

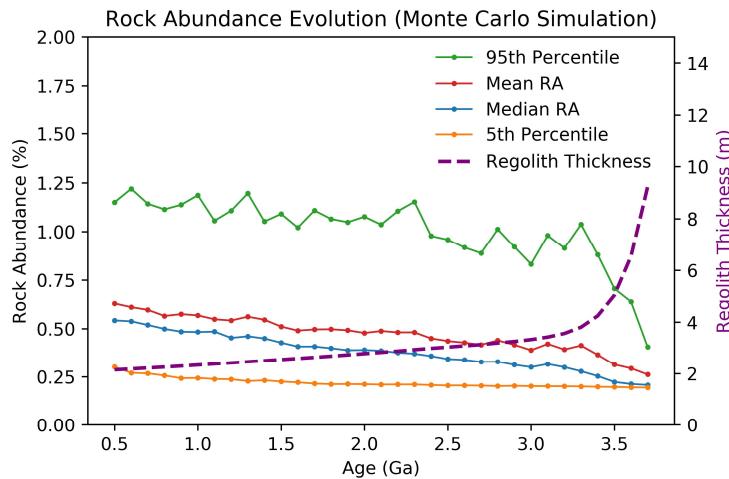
253 **4.2 Monte Carlo modeling of mare rock abundance**

254 To better understand how cratering and regolith thickness control the rock abundance, we
255 have written a simple Monte Carlo model of the rock excavation process to compare with our
256 observations (see also supporting Text S3 and repository code). For the growth of regolith, our
257 model follows Hirabayashi et al. (2018), which reproduces observed constraints on thickening of
258 the regolith on the mare well. Craters are generated at a rate defined by the NPF chronology
259 model and production function to cover the size range of D = 10 m to 800 m craters; additionally,
260 craters are generated in the D=5–10 m range using an impactor production function based on
261 Grün et al. (1985). To a first approximation, craters with diameter greater than ten times the local
262 thickness of the regolith are assumed to excavate rocks from the fractured bedrock at depth (e.g.,
263 Croft, 1980; Melosh, 1989).

264 The main free parameters of the model are (a) the proximal rock abundance from new
265 bedrock-excavating craters (rocks coming from bedrock), RA_{exc} ; (b) the proximal rock
266 abundance produced by new craters that do not reach bedrock (rocks coming from the existing
267 regolith volume), RA_{reg} ; and (c) the surface half-life of rocks, $\tau_{1/2}$. Values for these parameters
268 most consistent with observations were determined using a grid search. The mean, 5th, 50th, and
269 95th-percentiles of the model results and data were compared from 0.5 to 3.7 Ga. We also

270 artificially assign a non-zero regolith thickness at the start of the simulation, which improves the
 271 fit on young surfaces. This is needed because we do not simulate $D < 5$ m craters that contribute
 272 to the rapid growth of the initial regolith thickness. Variations in this artificial initial thickness
 273 parameter from 0 to 2-m do not affect the best-fit physical parameters, but do improve the
 274 quality of the best-fit.

275



276
 277 Figure 3. Best-fit Monte Carlo model of rock abundance evolution with time. The thicker regolith on
 278 earlier surfaces is why rock abundance is lower on surfaces >3 Ga.
 279

280 The best-fit model results are given in Fig. 3. This model has rock abundance from
 281 bedrock-excavating craters $RA_{exc}=0.08\pm0.02$, rock abundance from the regolith of RA_{reg}
 282 $=0.004\pm0.001$, and half-life of surface rocks $\tau_{1/2}=80\pm20$ Myr; error ranges were the fits with
 283 median symmetric accuracy of $<25\%$ (Morley et al., 2018). In a broad sense, this simplified
 284 Monte Carlo model does a reasonable job of reproducing the rock abundance evolution. With
 285 these best-fit parameters, rock exposure is most influenced by craters that form frequently but
 286 still reach beneath the regolith in the $D=10\text{--}25$ m size range; for the rockiest pixels (95th-
 287 percentile), larger crater sizes of $D=40\text{--}100$ m are more important.

288 The model observations provide strong constraints on the allowable parameters. For
 289 example, it is impossible to get a 5th-percentile rock abundance at 1 Ga that matches
 290 observations unless the half-life of surface rocks is small, $\tau_{1/2}<150$ Ma. Likewise, on old surfaces
 291 (>3.5 Ga), the 5th-percentile and median rock abundance are impossible to reproduce unless the
 292 fraction of rocks RA_{reg} coming from within the regolith is non-zero: otherwise the regolith is
 293 sufficiently thick on surfaces this old to prevent craters from replenishing surface rocks. This is

294 highly consistent with the interpretation of Elder et al. (2019), who note that a sharp, single
295 transition between regolith and underlying coherent rock cannot explain the observed rock
296 abundance trends.

297 We note there is a tradeoff between the rock abundance from bedrock-excavating craters,
298 RA_{exc} , and the rock half-life $\tau_{1/2}$ parameters: lower values of RA_{exc} need a larger $\tau_{1/2}$ to fit
299 observations, and larger values of RA_{exc} need a smaller $\tau_{1/2}$ to fit observations. There are
300 independent observational constraints that suggest $RA_{exc} \sim 0.04\text{--}0.10$ (e.g., South Ray + North
301 Ray; Ghent et al., 2014), although these are derived for larger craters than are important in this
302 model. Nonetheless, our results are consistent with these independent constraints. For the half-
303 life of surface rocks, our estimates are in close agreement with the values determined by
304 Basilevsky et al. (2013) of 40–80 Myr, though at the upper end of their range. Diviner data are
305 sensitive to slightly smaller rocks (~ 1 m) than the rocks visible in LRO (≥ 2 m). It is plausible
306 that there is a slight increase in the half-life of rocks as rock size decreases, although this is not
307 required by our results.

308 The qualitative ways that the model disagrees with the observations are twofold. First, we
309 do not observe the uptick in the 95th-percentile or mean rock abundance at young ages. As noted
310 earlier, this observation is based on only five bins, and thus may not be that robust. Second, the
311 model shows a steadier decline in rock abundance over time than the data, for any set of
312 parameters. This disagreement between the model and observations may be a consequence of the
313 limited precision of the rock abundance data, the limited fidelity of the neighborhood crater
314 frequency for discerning the local age, the simplifying assumptions of the model, or some
315 combination of these factors.

316 This Monte Carlo model is meant to be exploratory and is not intended to capture the full
317 details of cratering, regolith gardening, or rock excavation. A few simplifications of this model
318 are worth noting. Like Hirabayashi et al. (2018), the model neglects the lateral mobility of
319 regolith (and rocks) driven by topography or slopes. The model also does not include non-local
320 sources of rocks, such as from distal ejecta. Although rocks from distal ejecta likely occur on the
321 Moon, the average contribution of distal rocks is likely minor, since geochemical evidence
322 supports the interpretation that regolith is mostly locally derived (e.g., Papike et al., 1982). Our
323 measurements are also averaged over wide areas of maria of the same surface age. This is only

324 globally representative if rock excavation and destruction occurs at the same rate across the
325 surface. There is some evidence that some particular mare regions are unusually rocky for their
326 age (e.g., Humorum) (Hayne et al., 2017; Haber et al., 2018; Nypaver et al., 2022) that are worth
327 examining more closely in future work.

328 **4.3. Comparison with Earlier Work**

329 A past preliminary analysis of how rock abundance values compare to age was presented
330 by Haber et al. (2018). They used the mapped ages of units from crater counts by Hiesinger et al.
331 (2010), and also found a negative correlation between median rock abundance and age (see also
332 Figure S2). They also stated that the 95th-percentile RA value behaved similarly to the median.
333 Overall, their findings agree with our analysis, although our two analytical approaches differ in
334 some significant ways. The crater mapping underlying the Hiesinger et al. (2010) model ages
335 used by Haber et al. are better informed by geological assessment of unit boundaries than the
336 data we use from Fassett and Thomson (2014), but the Hiesinger et al. count areas for sampling
337 were relatively small, so ages on broader mare units are extrapolated. The neighborhood crater
338 frequency data used here are not extrapolated – there is a frequency and thus model age
339 calculated at every pixel – and thus cover more of the maria, but it does not separate individual
340 units as cleanly. The effects of this are mitigated by the way we bin large areas. Our strategy of
341 excluding regions near $D \geq 1$ km craters from the rock abundance analysis (though not from
342 contributing to neighborhood crater frequencies) also helps to focus the analysis on the
343 intercrater parts of the mare, where the background regolith development process is most
344 important compared to earlier work. Regardless, the agreement of the present results with earlier
345 work helps support the idea that an observable change in rock abundance occurs with unit age.

346 Xie et al. (2021) have used proxies for regolith thickness at several sites on the Moon
347 (anchored by the Apollo and Luna sites) to argue that the regolith thickened less quickly before
348 3.5 Ga than expected, which they suggest may imply a shift in the impactor population prior to
349 3.5 Ga. Our data and modeling do not reveal this slower-than-expected growth in regolith prior
350 to 3.5 Ga. Instead, we would argue that the rockiness we observe on mare units 3.5–3.7 Ga is
351 reconcilable with expected regolith growth with a consistent impactor population and the
352 expected higher impact flux. However, both Xie et al. and our measurements are indirectly
353 sensitive to regolith thickness. More in-situ measurements of regolith thickness, particularly

354 capable of accounting for its variability even on a single geologic unit, would be very valuable to
355 test these ideas. This could presumably be accomplished by a long-range lunar rover if outfitted
356 with appropriate geophysical instrumentation.

357 **4.4. Summary and conclusions**

358 Comparing Diviner rock abundance values to surface ages, we demonstrate that the rock
359 abundance on the maria decreases with neighborhood crater frequency and age in a significant
360 way. The rock abundance on maria units declines more quickly on surfaces older than 3.3 Ga,
361 consistent with thicker regolith due to the higher impact flux early in the Imbrian period.

362 We find a geologically short half-life of meter-scale surface rocks of 80 ± 20 Ma is most
363 consistent with our observations. This estimate also agrees with boulder lifetimes derived from
364 LROC observations of rocks in crater ejecta by Basilevsky et al. (2013). Because of this short
365 lifetime, our modeling suggests the rock abundance would be reduced below what is observed if
366 rocks were sourced only from bedrock. This supports a recent finding of Elder et al. (2019) that
367 re-excavation of rocks from within the regolith are important to the observed surface rock
368 abundances. Overall, these results provide valuable information about how rock populations on
369 the surface of the Moon evolve that can help understand landing sites that may be visited in the
370 next decade.

371

372 **Acknowledgments**

373 SV and MZ acknowledge funding from the NASA Lunar Data Analysis Program. We
374 thank two anonymous reviewers for constructive and clarifying suggestions that helped improve
375 the manuscript.

376

377 **Data and Model Availability**

378 Data and code to help reproduce the results of this paper have been released to Zenodo.
379 The code repository is available at doi:[10.5281/zenodo.6012585](https://doi.org/10.5281/zenodo.6012585) (also available at github,
380 <https://github.com/cfassett/MareRockAbundances>); the raw extracted data is in at archive with
381 doi: [10.5281/zenodo.6011671](https://doi.org/10.5281/zenodo.6011671).

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2

3 *Geophysical Research Letters*

4

Supporting Information for

5

6 **Rock abundance on the lunar mare on surfaces of different age: Implications for regolith
7 evolution and thickness**

8

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Contents of this file

22

Text S1 to S3

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25

26

27 **Introduction**

28 The supporting information includes additional model description and two additional figures
29 showing derived rock abundance observations.

30 **Text S1. Conversion of n(0.8-5km) Crater Frequencies to Model Age**

31 The crater measured in Fassett and Thomson (2015) were all between 800m and 5km.
32 Conversion of these to an equivalent model age based on the Neukum chronology can be
33 accomplished as follows, where capital N(X) represent the expected cumulative crater frequency
34 from diameter X to infinity:

35 $n(0.8-5\text{ km}) = N(0.8 \text{ km}) - N(5 \text{ km})$

36 The cumulative frequencies $N(0.8 \text{ km})$ and $N(5 \text{ km})$ can be calculated from the canonical
37 Neukum chronology function at $N(1 \text{ km})$

38 $N(1 \text{ km}) = 5.44e-14 [e^{6.93T} - 1] + 8.38e-4T$, where T is the model age in billions of years.

39 Scaling using the Neukum production function:

40 $N(0.8 \text{ km}) = 2.24468 * N(1 \text{ km})$

41 $N(5 \text{ km}) = 0.01384 * N(1 \text{ km})$

42 So $n(0.8-5\text{ km}) = 2.23084 * N(1 \text{ km})$.

43

44

45 **Text S2. Statistical significance**

46 We tested for the statistical significance of a trend in rock abundance as a function of
 47 neighborhood crater frequency with the Mann-Kendall test as implemented by *pyMannKendall*
 48 (Hussain and Mahmud, 2019). The table below shows the resulting parameters. There is a
 49 statistically significant trend in all of the summary statistics (median, average, 5th-percentile, and
 50 95th-percentile) as well as in a sub-sampled version of the raw data. The trends were strongest
 51 as measured by Kendall Tau-B in the average, median, and 95th percentile, consistent with
 52 qualitative examination of the data (e.g., Fig. 2a).

	Reject Null Hypothesis	Trend	p	Normalized Test Stat	Kendall Tau-b (larger negative number; stronger declining trend)
Median	True	Decreasing	3.55E-15	-7.868	-0.428
Average	True	Decreasing	0.00E+00	-10.164	-0.690
5th	True	Decreasing	3.40E-03	-2.929	-0.059
95th	True	Decreasing	2.22E-09	-5.981	-0.387
Raw Data/ subsampled*	True	Decreasing	0.00E+00	-54.783	-0.124

*For memory reasons, it was impossible to run the full 32million points in the dataset without subsampling. Every 400th point was selected as a subsample. Testing suggests tau_b does not change as a function of sampling, though more points result in a more significant test statistic, as expected.

53

54 *References:*

55 Hussain, Md. M. and Mahmud, I. (2019). pyMannKendall: a python package for non parametric
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58

59 **Text S3. Additional Monte Carlo Model Description**

60

61 More details of how the Monte Carlo model works are as follows (*model functions & variables in*
62 *parentheses*):

63 Iterations (nruns): Like all Monte Carlo models, this model is iterated a number of times,
64 *nruns*. The multiple trials are important because the sequence of craters is important to the
65 results. We used 1000 runs in all the results reported in this paper.

66 Space (domainpx/pxres): We used a single Diviner pixel as the relevant domain scale
67 (*domainpx* × *domainpx* = 236.9m × 236.9 m). This is divided into cells; we used *pxres* ~2m/px
68 resolution (118 cells × 118 cells).

69 Time/Age (endtimeMoon/modeltimestep/neqtime): For each Monte Carlo iteration, the
70 end of the model run is specified in physical units. We run the model from T=0 to *endtimeMoon*
71 =3.72 Ga in our calculations, and compare the data from 500 Ma to 3.72 Ga to cover the range
72 of the data we gathered. We do not plot the youngest times in Fig. 3 because the model
73 behavior in early times is particularly inaccurate because of the non-inclusion of <5m diameter
74 craters.

75 The time step in the model was defined as *modeltimestep*=50 Myr; experiments suggest
76 that this does not influence results. The expected number of craters in a given *modeltimestep* in
77 the domain remains constant at every timestep, though the exact number produced at a given
78 size depends on a draw from the Poisson distribution. The model timestep is equivalent to the
79 physical timestep only in the linear part of the NPF (Ivanov, 2001). For the exponential part of
80 the crater chronology, to produce the same number of craters, the equivalent physical timestep
81 is shorter to accommodate the much higher early flux. See the helper function *neqtime* in the
82 code repository in *modelparams.py*.

83 Tracked Quantities (lunarreg, lunarrockfa): Two quantities are tracked: the regolith
84 thickness (in meters), and rock abundance (meant to be conceptually equivalent to Diviner rock
85 abundance, so a fraction of the area covered by rocks (>=1m), not individual rocks).

86 Crater, regolith, and rock creation (drawrocks, drawreg): The size range used in the
87 model is from D=5m to 800m craters. The size and number of all the craters produced in a
88 timestep are determined from the NPF (with an extension below 10m using Grun et al. (1985) at
89 5m-10m), and these craters are added to the domain. Each crater's center position is drawn
90 randomly; the associated regolith added for each crater is calculated using Hirabayashi et al.
91 (2018)'s formulation. Inside the crater cavity, new regolith is only created if it is larger than what
92 exists already; outside, regolith (ejecta) is added to what already exists and declines with
93 distance from the rim as Dist^{-3} (e.g., Melosh, 1989).

94 Rock areal coverage is added with a heuristic depending first on what fraction *f* of the
95 new crater's interior location had regolith thickness greater than the excavation depth. The new
96 crater rock abundance is $f \times RA_{\text{exc}} + RA_{\text{reg}}$ in its interior and at its rim, with the excavated rock
97 abundance declining with distance from the rim as Dist^{-3} . RA_{exc} is meant to parameterize the
98 rock abundance produced from new bedrock-excavating craters (rocks sourced from bedrock);
99 RA_{exc} specifically is meant to represent the rock abundance of these zero-aged craters proximal

100 to their rim (i.e., in the model, this is treated as the area with the highest rock abundance and
101 the rock abundance then falls off with distance from the rim, as specified above). Added only
102 within one crater diameter, RA_{reg} is meant to conceptually reflect the addition of rocks from the
103 regolith volume (at a low rate): it happens whether the excavation depth is exceeded or not
104 (regardless of f). Note that, for simplicity, the rock abundance after new craters does not go
105 down. Using the maximum of the new and existing value does neglect the possibility that
106 excavation of regolith could occasionally reduce the rock abundance on surfaces via burial.

107 Rock destruction (*destroyrocks, halflife*): For every pixel, the areal coverage of rocks is
108 reduced exponentially at a *halflife* specified in the model parameters.

109

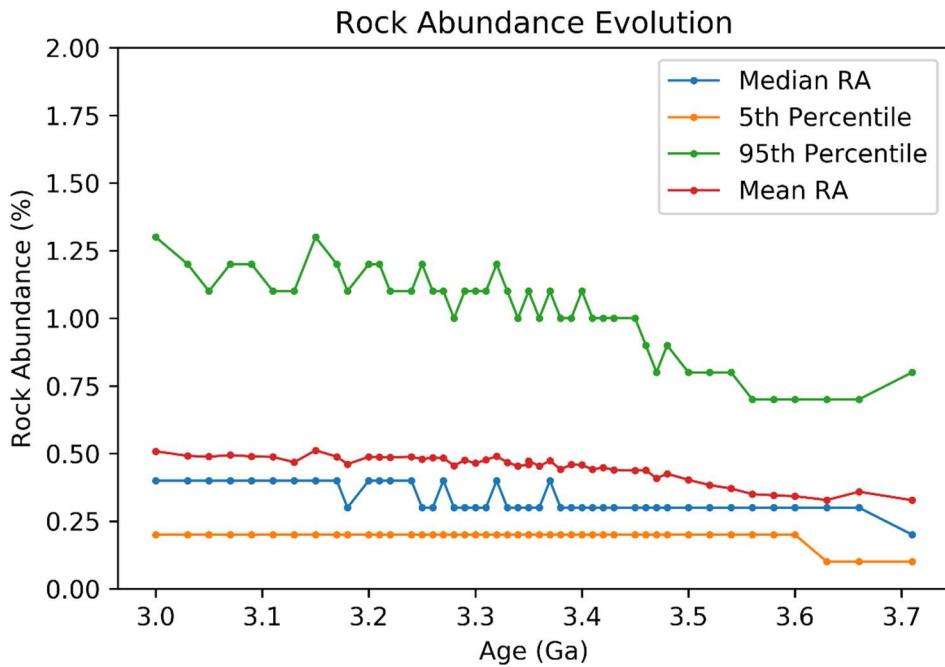
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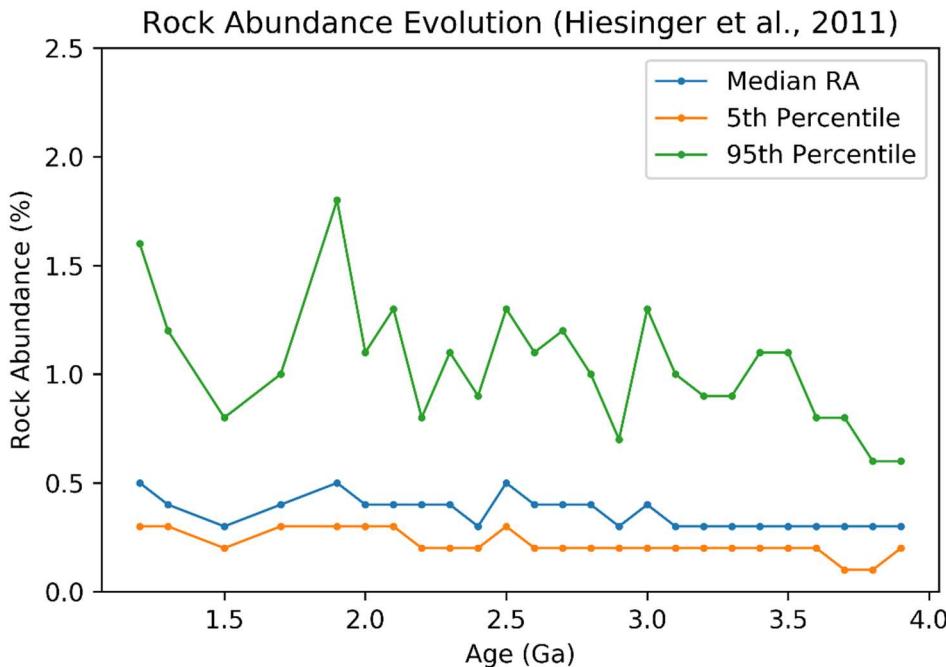
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121
Figure S1. Detail for the rock abundance evolution (from Fig. 2b) for the period before 3 Ga. The
122 trend in how rock abundance varies for surfaces of different age is more pronounced than later
123 in lunar history.
124

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126

127 **Figure S2.** Results for Diviner rock abundance (y-axis) vs Neukum model age in billions of years
 128 (x-axis) for units defined in Hiesinger et al., 2010. The Hiesinger unit boundaries used were from
 129 a digitization of the original Hiesinger et al. crater counts released by the LROC team. The bin
 130 sizes are defined in 100 Ma increments. The statistics are derived in a manner similar to Fig. 2.
 131 There is a general decline in the rock abundance, although less pronounced, than using
 132 neighborhood crater frequency to determine age.

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